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# **Beyond Lift-And-Shift:** **Optimizing Cloud Storage For IO Bound Workloads**

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# Agenda

## Use cases

- Database and Analytics IO
- Cloud deployment

## Architectural challenges

- Parallelism and low latency
- Scaling

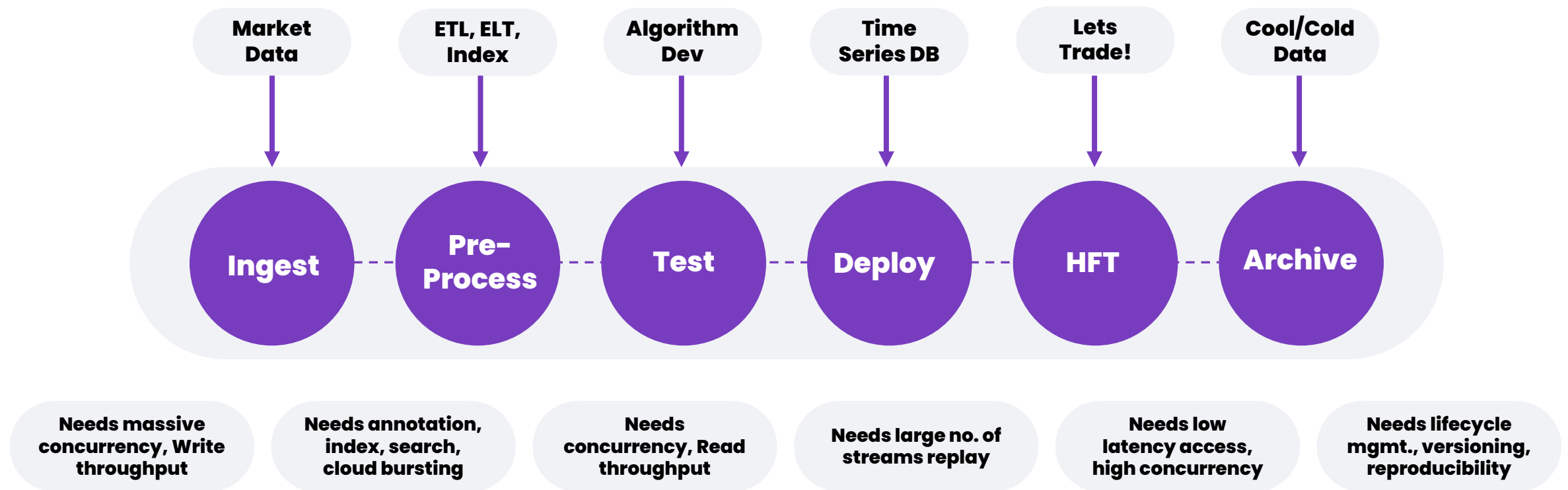
## MetaData Madness

- Appliance dedicated? Parallel but centralized MD?  
Distributed MD?

## WEKA Conclusions



# Use cases: WorkFlow and IO



# Use cases: Cloud usage



**ON**

**WITHIN**

**TO**

**WITH**

**BETWEEN**

Run Natively on  
the Cloud

Tier and Reduce  
Data Within a  
Cloud

Move or Backup  
to Clouds  
(Hybrid)

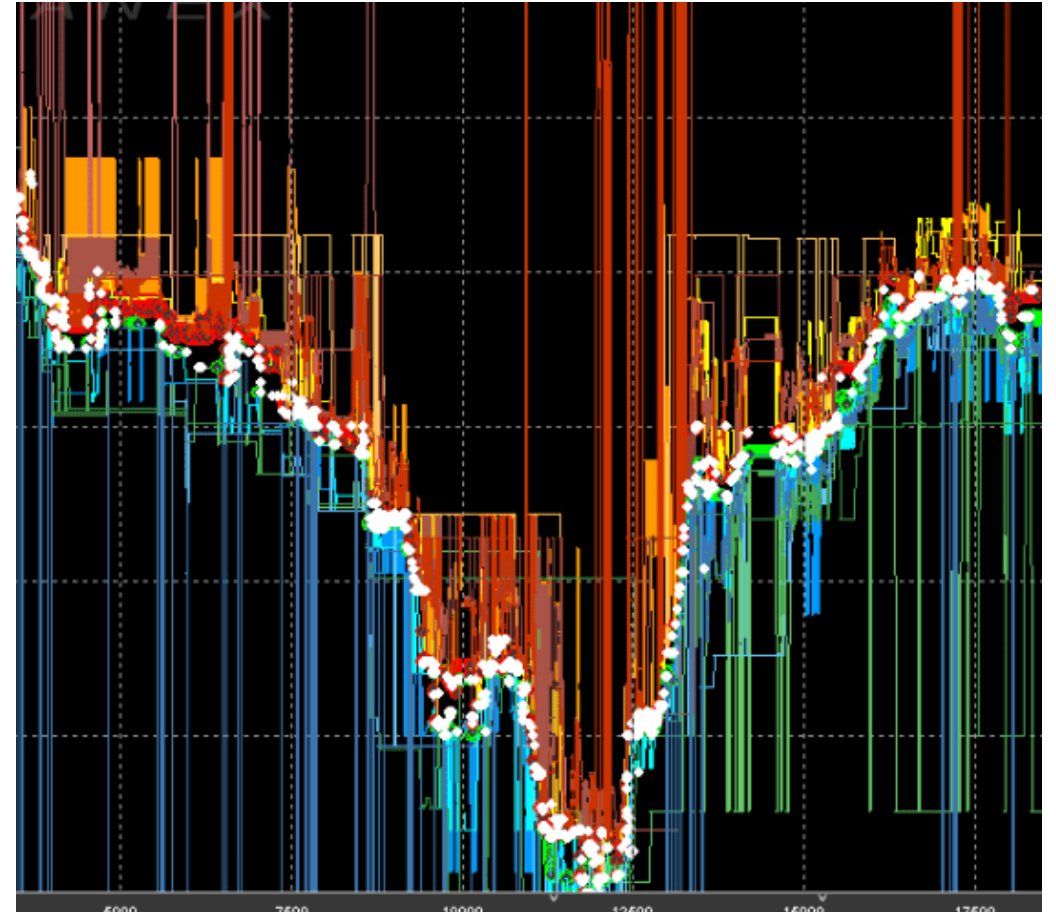
Use the Cloud  
for Data Tiering  
(Hybrid)

Migrate or DR  
Between Clouds

# HFT Challenges: It's All About Latency

## Trader's Tech Stack is a Major Competitive Advantage, Often its the Competitive Advantage

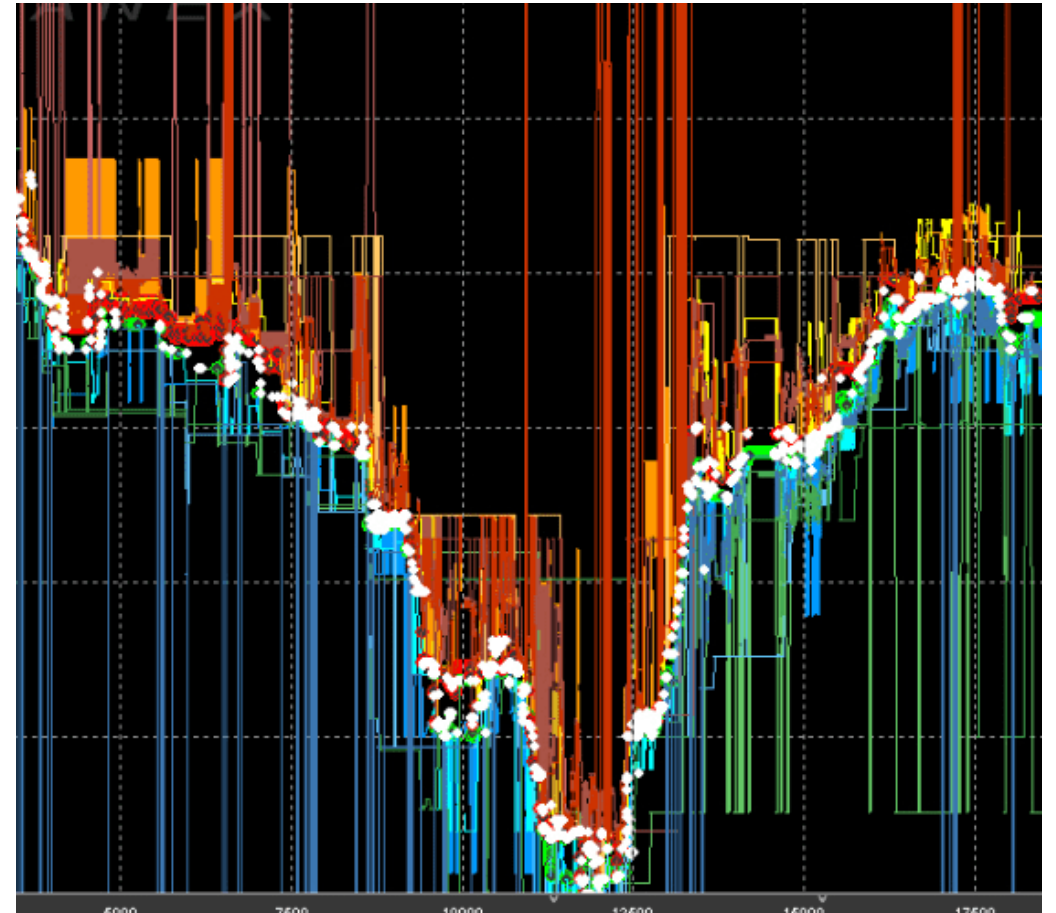
- High-Frequency Trading (HFT) slashes latency on market information to make profits before the competition
- In HFT, latency is often the only determinant for profit, or loss
- Not all algorithmic trading is high-frequency, but traders will still optimize their trading platform to execute any trade with low latency to get the best price execution



# HFT Storage in the cloud

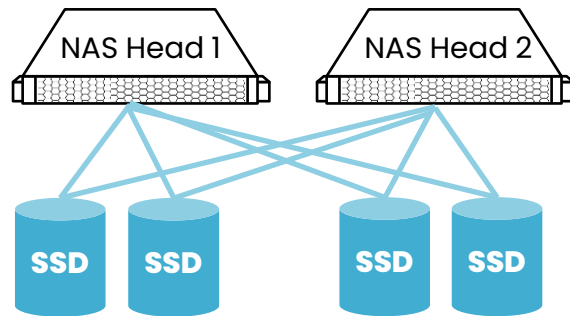
## What to choose

- Most HFT needs lots of random IO for the analytics powering the recommendation system/ Time Series DB. Think STAC-M3 types of access.
- Low latency requires fast instances
  - NVMe drives, Fast networking
- Single Availability zones and instance grouping help keep Instances geographically and logically close to reduce network hops unless additional reliability is needed.
- RAM is usually more critical for the trading clients than in the storage itself\*\*



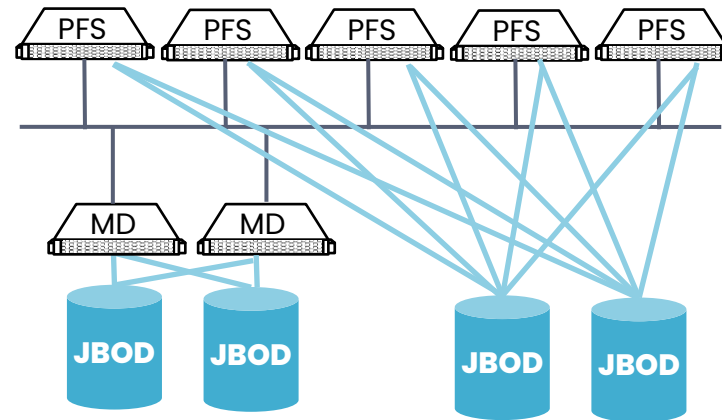
# Different Architectures: Metadata Madness!

## The Appliance



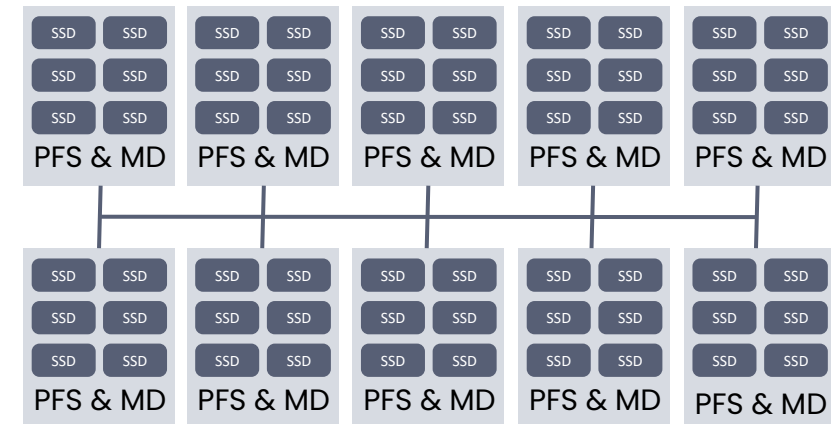
- NAS or SAN
- Limited to HA Pairs
- Metadata scale is limited to HA pair.
- Cloud Implementation recreates appliance in cloud even if it's software. May be co-located hardware.

## Parallel FS



- Specialized client (POSIX)
- Scale out
- Each FS uses dedicated MD resources
- Cloud Implementation may be co-located hardware
- Native cloud implementations are non-tunable

## Distributed Parallel

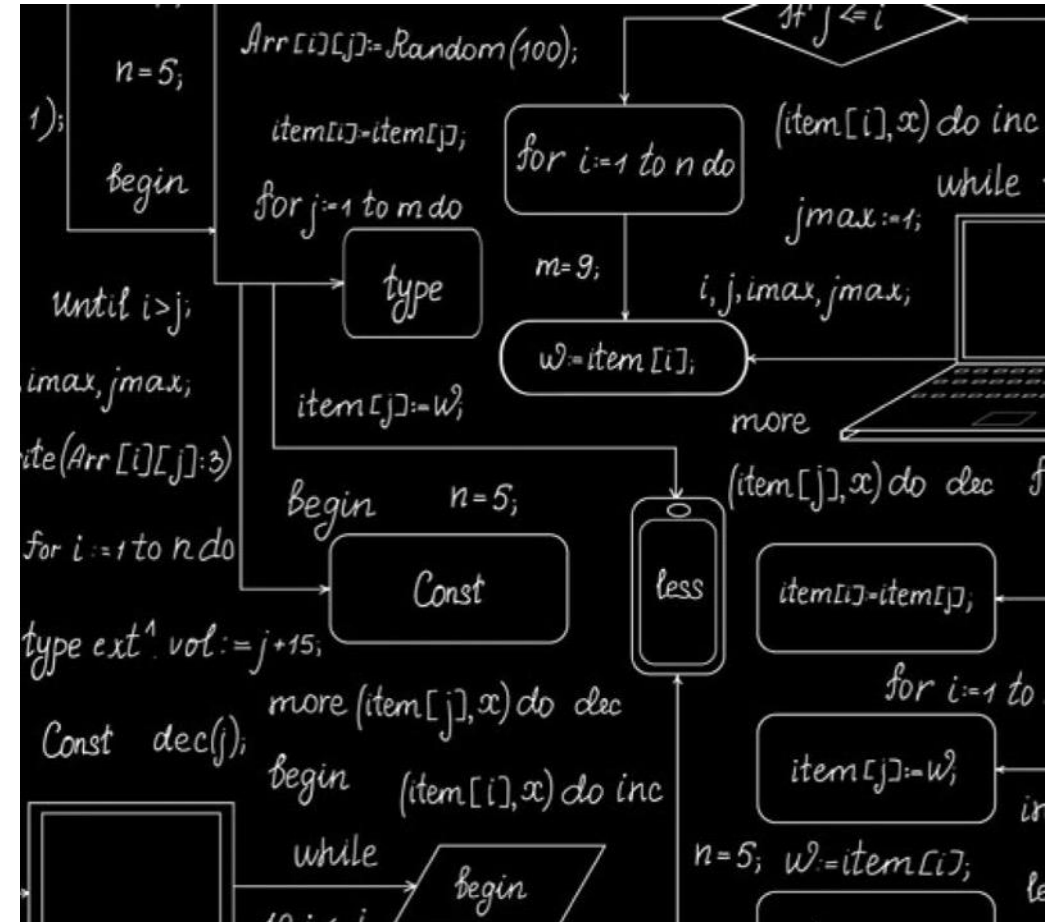


- Specialized client (POSIX)
- X86 based. Scale out or scale up
- Virtualized distributed MD in each server
- Cloud Implementation is Software running in instances
- Full SW implementation

# Algorithm Development: Data Feeds More Data

## Strains Infrastructure as Data Must be Continuously Fed to the Algorithm for Best Results

- A mix of data science, statistics, risk analysis and DevOps
- Algorithms are used for back-testing or experimenting against past data
  - Repeated to refine the algorithm
- Once results are verified, the algorithm is put in production
- Algorithm trading in the real-world markets will produce data that further feeds the algorithm backend

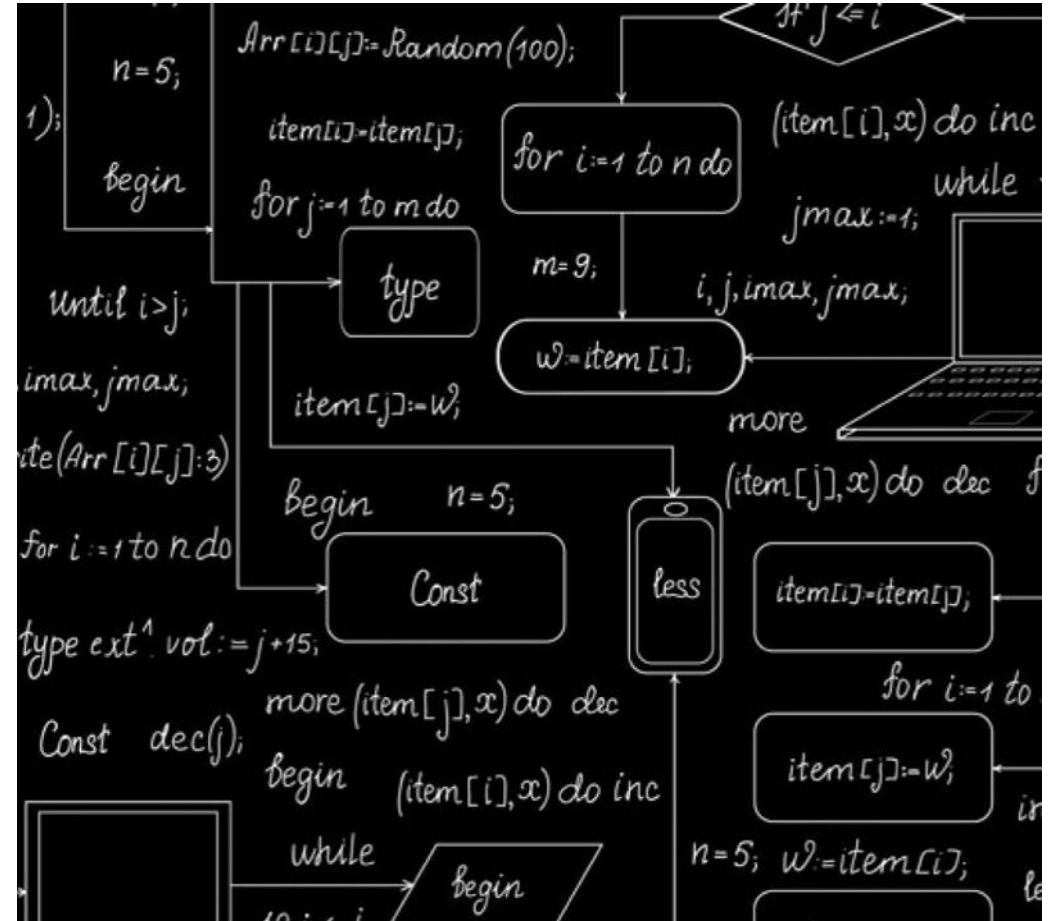




# Algorithm Development

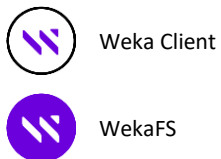
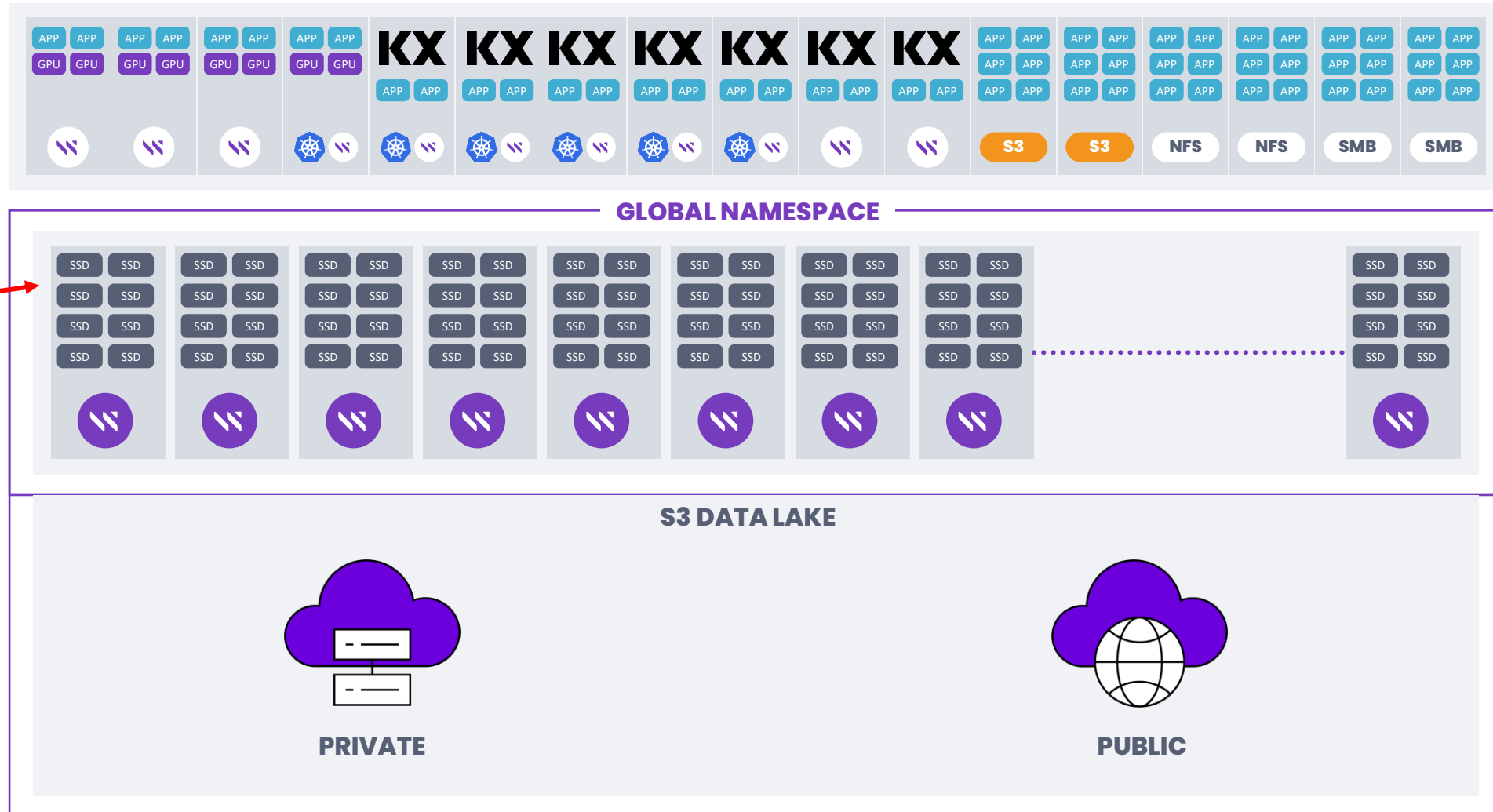
## What to Choose

- High number of developers and test running in parallel = Data Blender
- High performance in multiple dimensions is crucial. Storage needs to be flexible on types of IO
- Backtesting requires access to large amounts of historical data
- Object store access for lower cost for cool/cold data. NVMe becomes expensive at scale.
- Scale can be thousands of clients and 10s of petabytes of data (or more)



# Distributed Parallel: WekaFS

- Small 4K block size matches NVMe media 4K blocks to expose full performance of NVMe (others have 64K-1MB)
- Kernel bypass via DPDK, SPDK eliminate context switches, reduce kernel resources, pushes queuing towards zero
- All Metadata stays in the flash tier



# Conclusion: Don't be afraid of the cloud

- **Determine your workflow requirements**
  - High IOPS/small IO?
  - Big Throughput?
  - Data set sharing?
- **High performance is available!**
  - Low latency
  - High Concurrency
  - Faster Tick Analytics
- **Choose which cloud meets your needs**
  - AWS, GCP, Oracle, Azure
- **Integration with cloud object stores for cool/cold data**



# Q&A

# Thank You!

 @wekaio

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# Backup

# Testing in the cloud: SUT# KDB210507 in AWS

## 3 outright records on STAC-M3 Kananga

- WEKA was faster than other on-prem solutions as well

	STAC System Under Test #	WekaFSv3.10 in Kanaga benchmarks	WekaFSv3.10 in Antuco benchmarks
Lustre + Appliance	<i>#KDB200915</i>	Faster in 20 of 24 benchmarks	Faster in 4 of 17 benchmarks
Direct attached 10 servers with Optane	<i>#KDB200603</i>	Faster by 16 of 24 benchmarks	Faster in 9 of 17 benchmarks
All-Flash NAS	<i>#KDB200914</i>	All-Flash NAS did not submit Kananga benchmark	Faster in 15 of 17 benchmarks

# Multicloud Availability

- **Identical Code Base**
- **AWS**
  - CloudFormation scripts, full AWS API's, Autoscaling
  - Better cost/performance profile than FSx native services
- **GCP**
  - Terraform deployment, Autoscaling
  - Brings high performance storage to GCP
- **OCI**
  - Terraform deployment, Autoscaling
  - Integration with Oracle workloads
  - Insane performance: 2TB/s!
- **Integration with ALL 3 cloud object stores!**

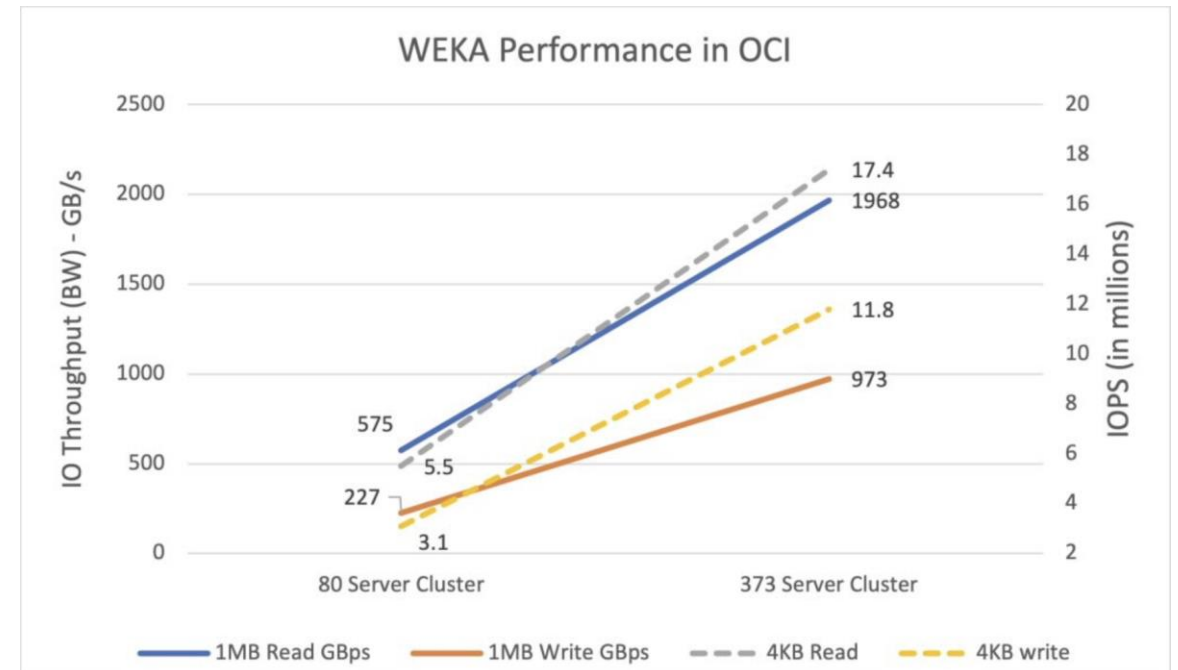




# WEKA on OCI Delivers 2 TB per Second Performance

## Maximum performance at cloud scale

- Run at petabyte scale in a high-performance file system
- NVMe SSDs for hot data and object storage for warm or cold data
- High-performance computing (HPC) bare metal Compute shape (BM.Optimized3.36)
- 100-Gbps RDMA over converged ethernet (RoCEv2) and 3.8 TB of local NVMe SSD



<https://blogs.oracle.com/cloud-infrastructure/post/weka-on-oracle-cloud-infrastructure-delivers-2-terabytes-per-second-performance>